

Minimal Overlap Observed in Cell Types Associated with Parkinson's Disease GWAS

Dhilip Raman, July 31st, 2023 PID 2740092 Complex Trait Genetics (CTG)

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Abstract

Functional interpretation of genome-wide association studies has been previously focused on identifying causal SNPs, but identifying cell types associated with a disease trait can also be important for unraveling disease pathology. The FUMA platform (fuma.ctglab.nl) has provided the Cell Type tool which maps cell types to a single GWAS using gene expression profiles of clustered cell types; in this study, we aim to extend the visualization of a disease trait's association with cell types by comparing multiple GWAS of the same trait in parallel. To this end, we provide a multi-GWAS exploratory tool in RShiny which compares the overlap in cell types, genes, and genomic risk loci associated with multiple GWAS of the same trait. We demonstrate the usage of this tool by exploring the four most relevant Parkinson's Disease GWAS from the last fifteen years. In our study, we find that the cell types associated with Parkinson's disease are not concordant across all PD GWAS. However, the cell types and genes associated with Parkinson's disease in our study do validate previous research, particularly the association of monoaminergic neurons and oligodendrocyte glia cell types with Parkinson's disease pathology.

Introduction

Genome-wide association studies (GWAS) have been an important tool for_identifying genetic variants for many traits and diseases. Disease traits are investigated in GWAS by genotyping large populations and calculating differences in SNP (single-nucleotide polymorphism) frequencies between individuals with the trait and individuals without the trait (Uffelman et al, 2021). In recent years, GWAS have become increasingly more powerful, as meta-analysis methods and cheaper genotyping technology have allowed newer GWAS to include over 1 million participants (Loos, 2019). With a sufficiently powerful GWAS, all SNP variants associated with a trait of interest could potentially be identified (Wang & Chatterjee, 2017); however, further functional annotation of GWAS to specify a SNP variant as a causal variant for a disease requires new approaches (Wu et al, 2017).

In the past, researchers have investigated which cell types are important for a trait or disease to functionally follow up a GWAS result. To link GWAS to cell types, researchers usually identify gene sets or gene properties that define certain cell types based on single cell RNA (scRNA) sequencing data; these datasets quantify the gene expression in each cell. The gene expression profiles of individual cells is subsequently used to cluster cells into cell types. Researchers can then test whether GWAS signal is enriched for the genes belonging specifically to one cell type.

In a previous study, an atlas of 4,155 GWASs was cataloged (Watanabe et al, 2019); this resource allows researchers to view risk SNPs identified by each GWAS and but does not yet inform on which cell types are associated with these SNPs. The FUMA platform also provides a tool for mapping cell type to GWAS signals (Watanabe et al, 2017), but comparing multiple GWAS and their associated cell types remains difficult. Thus, our project aims to produce a tool that visualizes FUMA's functional analysis for a trait, by comparing FUMA results for multiple GWAS of the same disease trait to identify genes or cell types that are implicated repeatedly.

Cellular gene expression is measured in scRNA sequencing studies, which FUMA maps to the genetic expression profile observed for a disease trait investigated in a GWAS. Single-cell RNA sequencing datasets will be selected for this analysis based on their most specific level of annotation: many popular scRNA sequencing databases for brain cell types, such as the Allen Brain human and mouse datasets, have been updated multiple times to have more specific labeling of cell types.

To illustrate our approach for comparing cell types associated with a disease trait, our study will focus on Parkinson's disease. While the phenotype for Parkinson's disease is clear, there is more to explore about which genetic variants drive Parkinson's disease and the relevant cell types that could be targeted for treatment. Parkinson's disease (PD) pathology is characterized by a progressive loss of autonomous motor function, which is linked to loss of dopaminergic (DA) neurons in DA-rich regions of the brain such as the midbrain (Kouli, *Parkinson's Disease*). In the brain-first progression of PD, alpha-synuclein aggregates, known as Lewy bodies, appear first in the medulla and olfactory bulb and contribute to toxicity-driven degeneration of DA neurons (Yamasaki et al, 2022). Parkinson's is a progressive disease, with alpha-synuclein driven degeneration spreading to different regions of the brain slowly over the course of decades. PD progression is described with Braak staging: 5 stages of worsening degeneration as the disease spreads from the substantia nigra to deeper regions of the midbrain and eventually to other brain regions and the body (Dickson et al, 2010). Recent analysis of PD patient-derived induced pluripotent stem cells (iPSCs) has confirmed the accuracy of Braak staging and the implicated brain regions (Fernandes et al, 2020). However, little is known about cell types other than dopaminergic neurons that might contribute to the PD progression described by Braak stages.

Since treatment for Parkinson's disease has been primarily focused on restoring healthy dopamine signaling (Armstrong et al, 2020), further study into specific neuron and glia cell types implicated in PD pathology is necessary for improved targeted treatment of alpha-synuclein-driven degeneration. In this study, we will address our research question: which genomic risk loci, genes, and cell types repeatedly emerge from PD GWAS? For this study, we are assuming that the genomic loci, genes, and cell types that map most concordantly to PD GWAS are also the most associated with PD pathology. Previous cell-type mapping (Bryois et al, 2019) to GWAS for Parkinson's disease has implicated oligodendrocytes in addition to the monoaminergic class of neurons (which contains dopaminergic neurons), so in our study we hypothesize that glia will be associated with some of the PD GWAS we investigate; however, since our GWAS vary in number of participants and reference populations, we might not see a concordant implication of cell types across all PD GWAS.

In this study, we first tested if cell types identified from multiple PD GWAS and different scRNAseq datasets converge. Next, we developed a tool to explore genomic risk loci, genes, and cell types that are associated with PD, with interactive elements that allow researchers to navigate cell types prioritized by different PD GWAS and scRNAseq datasets. This tool is aimed to help future researchers catalog which cell types are relevant to their disease of choice (and

potentially to other traits of interest). The pipeline for visualizing this overlap will be demonstrated using GWAS for Parkinson's disease, but we aim to extend this pipeline to other disease traits in the future.

Methods

To access GWAS summary statistics for Parkinson's disease (PD), the GWAS Atlas platform was utilized (https://atlas.ctglab.nl/). Nine studies are available on the GWAS Atlas for non-familial PD; we selected the four latest GWAS to analyze, excluding the five older studies for having too small of a population genotyped which results in no genome-wide significant SNPs, or for having incomplete data. The four GWAS selected (Supplementary Table 1) were Nalls et al (2019), Blauwendraat et al (2019), Pankratz et al (2012), and Simon et al (2009).

Pre-processing of GWAS summary statistics entailed two steps. First, SNP uniqIDs were split into separate columns of chromosome and position for each SNP. FUMA'S SNP2GENE tool requires these columns to be separate in order to recognize chromosome and position data. Next, pre-processed summary statistics were validated by checking the length of the file, and ensuring that it matched the line count for the same GWAS on the GWAS Atlas.

Using the SNP2GENE results, we first assembled a comparison of the overlap between genomic risk loci identified for Parkinson's disease. SNP2GENE identifies genomic risk loci using multiple "LD blocks", regions with largely self-contained linkage disequilibrium which result in associated gene expression. LD blocks vary widely in size, but within the LD block there is expected to be multiple candidate SNPs that are associated with a select few "lead SNPs" which potentially drive the phenotype (Watanabe et al, 2019). Since LD blocks can be of any size, we defined overlap in genomic risk loci as any genomic risk loci that overlap at any location within start to end. This comparison of genomic risk loci was plotted using an Upset diagram and overlapping loci were identified.

We next compared the nearest genes associated with significant SNPs. These genes are selected simply as those in the closest proximity to the SNPs that fall within the genome-wide significance threshold of 5*10^-8 (Watanabe et al, 2019). We compared overlap between GWAS to identify nearest genes that repeatedly emerge in association with PD. Overlap in nearest genes was plotted using an Upset diagram and overlapping genes were listed.

Lastly, we compared cell type associations with Parkinson's disease. Gene expression at a single-cell resolution is used to find cell types which express genes associated with GWAS traits. Mapping of GWAS to cell types is performed using MAGMA gene-property analysis, which relies on a multiple regression approach (Watanabe et al, 2019); this is implemented in FUMA's Cell Type tool. We used 31 scRNAseq datasets for cell-type mapping (Supplementary Table 3), which each capture different regions of the brain. We compared overlap in cell types associated with PD and plotted them using an Upset diagram. Additionally, we compared cell types identified in this analysis with PD cell types identified in previous studies and we also

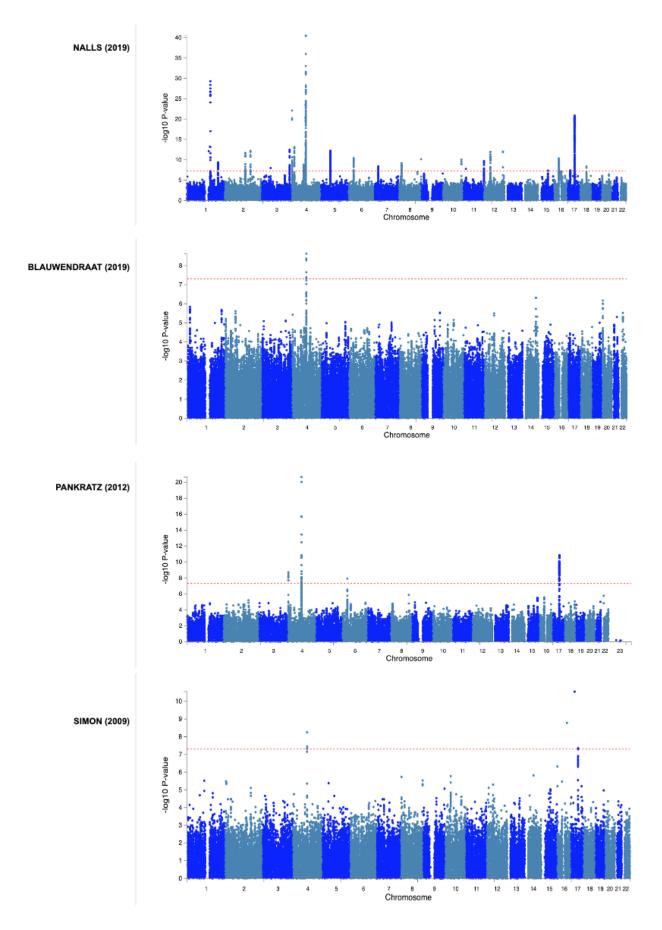


Figure 1: Manhattan Plots of the four Parkinson's disease GWAS selected for this study. SNPs are plotted as dots in the Manhattan plot. Chromosomal position is shown on the x-axis, and log-adjusted P-values are shown on the y-axis. The genome-wide significance threshold of 5*10^-8 is displayed as a dotted red line; all significant SNPs fall above this significance threshold.

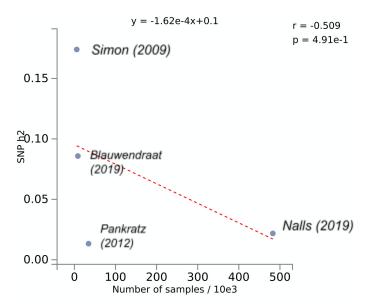


Figure 2: Scatterplots of the four Parkinson's disease GWAS selected for this study describing the relationship between number of participants in the study (showing the size of the study) versus the heritability (h^2) of Parkinson's disease captured by this study. Number of participants is shown on the x-axis, and heritability is shown on the y-axis. Each GWAS's name is listed on the plot.

further investigated whether any cell types from unique brain regions (based on the scRNAseq dataset) were identified.

To simplify these comparison tests, we created an application using RShiny to compare overlap in genomic risk loci, nearest genes, and cell types between multiple GWAS for the same trait. The goal of this app is to supplement FUMA results with a comparison tool, so that the user can download their FUMA results and use our pipeline to aggregate their results for multiple GWAS and easily visualize the comparison between GWAS.

The key feature of this tool is the identification of cell types that repeatedly emerge in association with PD. To this end, the first function of our Shiny app is a barplot that lists cell types and their p-values, denoting the significance of their association with PD. Using this barplot, the user can identify the names of cell types that are considered significant for PD. The user could also observe certain groups of cell types that repeatedly emerge, such as glial cells or excitatory neurons.

Finally, our app also describes the pipeline the user will need to follow in order to compare multiple GWAS. Users must first download their GWAS summary statistics from the GWAS Atlas, then use our pre-processing script to prepare their summary statistics, then run their summary statistics on SNP2GENE and Cell type on the FUMA platform. They can then use our Shiny script to aggregate the FUMA results and visualize them in the Shiny app.

Results

Across PD GWAS, there is a concordant identification of significant SNPs within chromosome 4, and some repeated identification of significant SNPs in chromosome 17 (Fig. 1). Nalls et al (2019) uniquely identifies multiple significant SNPs in chromosome 1 risk loci. However, although Nalls et al (2019) utilizes the highest number of samples, it also captures a very low heritability when compared to other PD GWAS (Fig. 2).

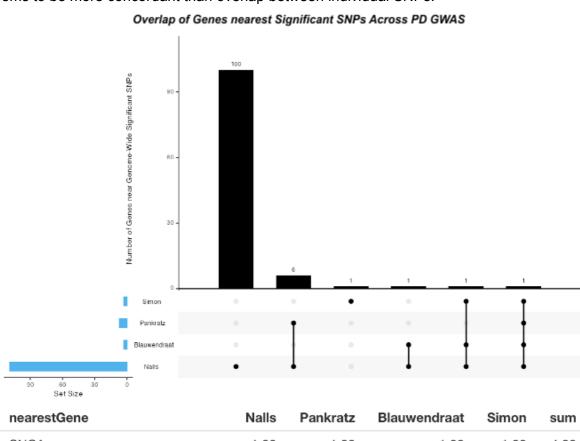
We also tested if genomic risk loci overlapped and found only one genomic risk locus in chromosome 4 which was identified by more than one GWAS (Fig. 3). A total of 30 genomic risk loci were identified between all four Parkinson's disease GWAS, located primarily in chromosomes 1, 4, and 17 (Supplementary Table 2). Of these 30 risk loci, 29 were unique to GWAS while one risk locus containing the SNP with rsID 356203 (located within chromosome 4)

was identified by both Blauwendraat et al (2019) and Nalls et al (2019). Overall, there is little overlap of genomic risk loci between Parkinson's disease GWAS, and investigating the locus that did repeatedly emerge did not result in a novel identification of SNPs associated with PD.

Overlap of Genomic Risk Loci Across PD GWAS 25 23 20 Number of Genomic Risk Loci Pankratz Set Size rsID Nalls **Pankratz** Simon Blauwendraat sum rs356203 1 0 0 2

Figure 3: Upset diagram of the overlap in genomic risk loci identified by the four Parkinson's disease GWAS. Number of genomic risk loci is shown on the y-axis, and GWAS is shown on the x-axis. Genomic risk loci identified by multiple GWAS appear not as singular dots, but as a line connecting multiple dots.

There is a much stronger overlap between genes nearest to SNPs significant for PD, with approximately forty genes in chromosome 4 and 17 being implicated by multiple GWAS (Fig. 4). SNCA, a gene that codes for the protein alpha-synuclein, is implicated by all four GWAS. Other genes that emerge from at least 2 GWAS include genes that code for retinal ganglion cells, tau phosphorylation genes, and corticotropin-releasing hormone genes (Fig. 4). A total of 110 genes were identified as those nearest to significant SNPs; of these 110 genes, 100 were uniquely identified by Nalls et al (2019), 1 was uniquely identified by Simon et al (2009), and 9 genes were identified from multiple GWAS. Therefore, overlap between PD GWAS at the gene level seems to be more concordant than overlap between individual SNPs.



nearestGene	Nalls	Pankratz	Blauwendraat	Simon	sum
SNCA	1.00	1.00	1.00	1.00	4.00
RP11-115D19.1	1.00	0.00	1.00	1.00	3.00
CRHR1	1.00	1.00	0.00	0.00	2.00
CRHR1-IT1:CRHR1:RP11-105N13.4	1.00	1.00	0.00	0.00	2.00
CRHR1:RP11-105N13.4	1.00	1.00	0.00	0.00	2.00
KANSL1	1.00	1.00	0.00	0.00	2.00
MAPT	1.00	1.00	0.00	0.00	2.00
MAPT-AS1	1.00	1.00	0.00	0.00	2.00
RP11-67M1.1	1.00	0.00	1.00	0.00	2.00

Figure 4: Upset diagram of the overlap in genes nearest significant SNPs identified by the four Parkinson's disease GWAS. Number of genes is shown on the y-axis, and GWAS is shown on the x-axis. Genes identified by multiple GWAS appear not as singular dots, but as a line connecting multiple dots.

Next, we investigated overlap between cell types associated with PD. We first identified all cell types that were considered significant after multiple testing correction with Bonferroni. The number of cell types within each scRNAseq dataset was used as the number of tests within multiple testing correction; the resulting value was called "P.adj.pds", which stands for P-adjusted within the scRNAseq dataset. We also applied a log10 transformation and ranked the cell types to identify those that were most associated with PD GWAS signal (Fig. 5). We identified a total of 20 significant cell types (Table 1), but each of these cell types were identified by a single GWAS. There is no overlap in significant cell types from multiple GWAS (Fig. 6). However, there is repeated implication of the same cell type groups. From different GWAS, excitatory neurons emerge repeatedly (vGLUT neurons, NF cortical neurons, Gad2 neurons). Oligodendrocyte, OPCs (oligodendrocyte precursors), and radial glia also appear as significant cell types but are only uniquely identified by Blauwendraat.

Dataset	Cell_type	logP.adj.pds	GWAS	BETA_STD	SE \$	P.adj.pds 🏺
Linnarsson_GSE59739_Mouse_Dorsal_root_ganglion_level3	NF3	2.350178	Nalls	0.079796	0.0094077	0.00446501
DropViz_ENT_level2	Neuron_Gad1Gad2_Dlk1.4_4	2.028348	Nalls	0.055650	0.0058997	0.00936810
Linnarsson_GSE74672_Mouse_Hypothalamus_Neurons_level2	Vglut2_11	2.002269	Nalls	0.066240	0.0197790	0.00994790
Allen_Mouse_LGd2_level3	vLGN_Inh_Rspo2	1.819956	Simon	0.086667	0.0103590	0.01513714
Linnarsson_GSE74672_Mouse_Hypothalamus_Neurons_level2	Vglut2_12_Mgat4b	1.733684	Nalls	0.046355	0.0212670	0.01846360
Allen_Mouse_LGd_level2	Gad2_Rspo2	1.727001	Simon	0.076191	0.0114280	0.01874990
GSE168408_Human_Prefrontal_Cortex_level3_Childhood	L5.6_THEMIS_CNR1	1.671238	Pankratz	0.078930	0.0367500	0.02131875
Linnarsson_GSE59739_Mouse_Dorsal_root_ganglion_level3	NF2	1.609446	Nalls	0.089371	0.0134910	0.02457840
PsychENCODE_Adult	Ex5b	1.591845	Pankratz	0.097402	0.0952280	0.02559500
GSE168408_Human_Prefrontal_Cortex_level3_Adolescence	L5.6_THEMIS_CNR1	1.526423	Pankratz	0.078228	0.0237320	0.02975616
DropViz_SN_level2	Neuron_Th_Grin2c.4_9	1.522063	Nalls	0.061436	0.0068175	0.03005637
GSE168408_Human_Prefrontal_Cortex_level3_Adolescence	CCK_SYT6	1.519529	Simon	0.069163	0.0345340	0.03023232
Linnarsson_GSE104323_Mouse_Dentate_gyrus	RGL	1.517561	Blauwendraat	0.041087	0.0978210	0.03036960
DropViz_GP_level2	Neuron_Gad1Gad2_Satb1.2_6	1.459557	Nalls	0.065556	0.0072735	0.03470907
DropViz_TH_level2	Oligodendrocyte_Trf_Ndrg2.9_1	1.443269	Blauwendraat	0.039579	0.0048816	0.03603552
Linnarsson_MouseBrainAtlas_level6_rank4	Chmn	1.430773	Nalls	0.048542	0.0287470	0.03708744
Allen_Mouse_LGd_level2	Slc17a6	1.383957	Nalls	0.056702	0.0091758	0.04130880
Allen_Human_LGN_level1	Non.Neuronal	1.374955	Blauwendraat	0.030580	0.0091413	0.04217400
GSE104276_Human_Prefrontal_cortex_per_ages	GW23_OPC	1.315600	Blauwendraat	0.055607	0.0102820	0.04835040
Linnarsson_GSE95752_Mouse_Dentate_gyrus	Radial_Glia_like	1.303678	Blauwendraat	0.032338	0.0616330	0.04969610

Table 1: This table lists all the cell types that are significant for Parkinson's disease, identified from our four Parkinson's disease GWAS. P.adj.pds refers to p-value corrected with Bonferroni within scRNAseq dataset, logP.adj.pds refers to the log-adjusted version of P.adj.pds, BETA_STD refers to the standardized beta, and SE refers to the standard error.

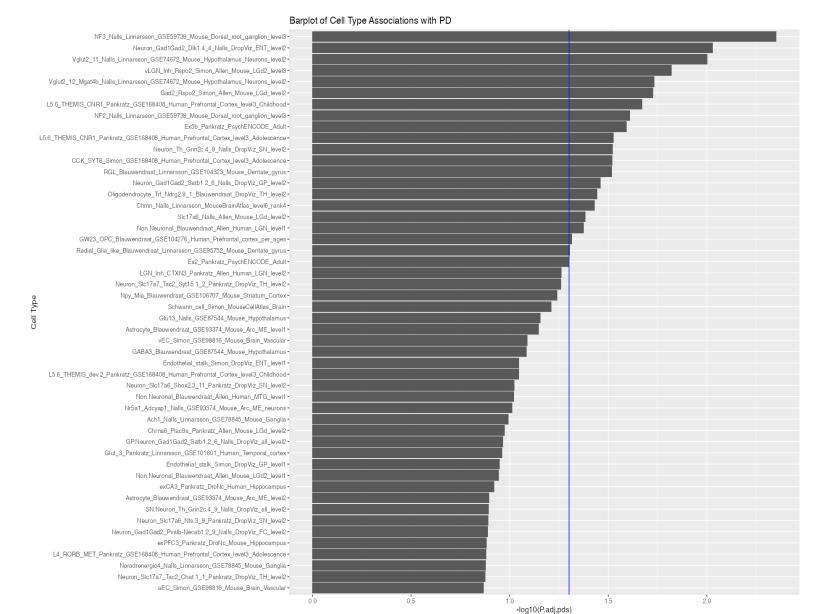


Figure 5: Barplot of cell types identified by the four Parkinson's disease GWAS. Cell type name and the GWAS and dataset it originates from is shown on the y-axis, and log-adjusted p-values are shown on the x-axis. The genome-wide significance threshold of 5*10^-8 is displayed as a blue dotted line. All significant cell types fall to the right of this significance threshold.

To visualize our results, the Shiny tool used in this project creates multiple Upset diagrams describing overlap, a barplot to list significant cell types, and a table containing all significant cell types. The user interface for this Shiny app is shown below (Fig. 7).

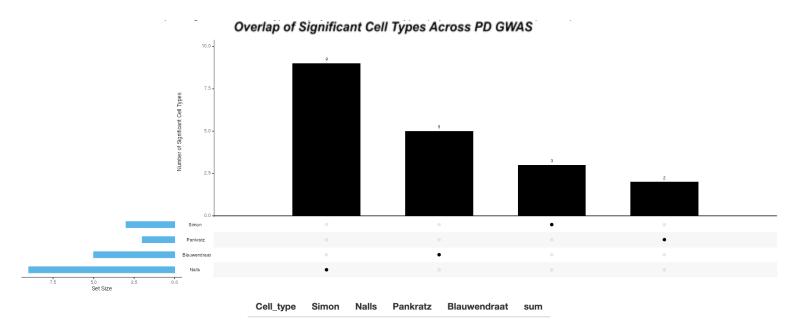


Figure 6: Upset diagram of the overlap in significant cell types identified by the four Parkinson's disease GWAS. Number of genes is shown on the y-axis, and GWAS is shown on the x-axis. Cell types identified by multiple GWAS appear not as singular dots, but as a line connecting multiple dots. In our study, there were no overlapping cell types identified.

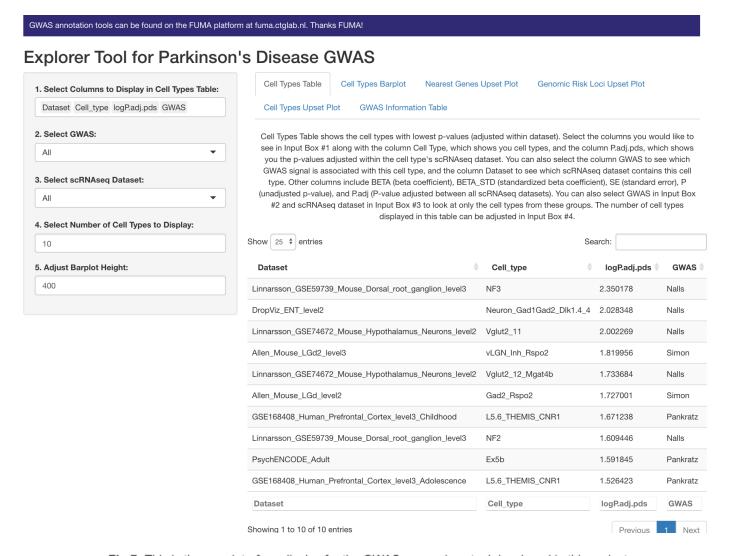


Fig 7: This is the user interface display for the GWAS comparison tool developed in this project using the Shiny package for R. This GWAS comparison tool displays overlap in genomic risk loci, nearest genes, and cell types, across multiple PD GWAS.

Discussion

In this project, we developed a tool to investigate genomic loci and cell types associated with PD. Our goal was to identify risk loci and cell types that repeatedly emerged from functional follow-up analysis of PD GWAS. However, our results suggest that PD GWAS demonstrate little concordance of genomic risk loci or of cell types, largely due to a lack of robust PD GWAS to compare. However, there is a moderate overlap of PD-associated genes. Genes nearest significant SNPs associated strongly with cell type functions (such as corticotropin release or tau phosphorylation) that have been shown to play a large role in PD (Zhang et al, 2018; Rissman 2007).

Genomic risk loci identified for PD GWAS did not show overlap across GWAS, but the risk loci identified did point to some interesting results. Most of the risk loci identified were in chromosome 4 and chromosome 17, and were mostly linked with the alpha-synuclein gene SNCA in chromosome 4 or the tau phosphorylation gene MAPT in chromosome 17 (Supplementary Table 2). Nalls et al (2019) identified 1 unique risk loci in chromosome 1 (Supplementary Table 2); this loci could be linked with the PARK family of genes. PARK genes have been studied for their link with monogenic parkinsonism; PARK mutations are linked with early-onset parkinsonism symptoms such as motor and speech deficits (Bonifati et al, 2013). However, chromosome 1 risk loci were only identified in Nalls et al (2019) and therefore lack the validity of association with PD that chromosome 4 and chromosome 17 risk loci have.

Across PD GWAS there was limited overlap in identified cell types. While PD maps poorly to single-cell expression, certain brain regions seem to be repeatedly associated with PD. This is shown in the multiple significant cell types found in the hypothalamus and dorsal root ganglia, in both mouse and human scRNAseq datasets (Table 1). These brain regions are strongly implicated in Parkinson's disease pathology (Kouli, 2018). Furthermore, the genes we identified (Fig. 4) as associated with PD (such as RP11 genes which code for retinal ganglion cells in the optic bulb) are also validated by PD pathology; the brain-first pathology of PD is also believed to originate in the optic bulb. From our results we can conclude there is more overlap between PD GWAS at the brain region level rather than at the cell type level.

However, identifying overlap in cell types is further complicated by the convention of cell type naming in scRNAseq datasets and the lack of standard naming across different datasets. Usage of multiple scRNAseq datasets in comparison is highly impacted by a lack of naming standards (Crow et al, 2018). In this study, multiple concordant cell types had to be renamed because of spelling differences and acronym usage. Furthermore, cell type annotations are highly specific, differentiating cell types with specific protein expression (i.e. MAPT vs MAPT-AS1 cells). Therefore, naming of cell types seems to play a very large role in limiting our identification of concordant cell types. In this study, we did attempt to standardize cell names within our datasets, but it remains a complicated process.

This study does identify some cell type associations with PD that could be further investigated. First, glial cells are weakly associated to PD when mapped with GWAS (Table 1), unlike with in

vivo studies of PD (Smajic et al, 2022), suggesting that glial cells populations may play less of a causal role in PD and instead are reactive to the initial disease. It has been shown though that glia cell death exacerbates the disease (Olsen et al, 2021). Astrocytes and microglia have been previously linked to PD, but were largely absent from our list of significant cell types (Copas et al, 2021). Instead, our results show that NF cortical cells and OPCs were implicated repeatedly as cell types significant for PD, and both cell types play a role in the myelination of nerves (Dean et al, 2016). Yet PD pathology shows little effect on myelination; progressive demyelination does not present itself in PD as it does in other neurodegenerative diseases (Le et al, 2023).

As a tauopathy, PD may also have a more complicated cell pathology because of multiple types of cell-to-cell transmission of tau (Gibbons et al, 2019). Tauopathies result in highly heterogeneous pathways of disease progression at the cellular level (Chung et al, 2021), so while brain regions affected by PD are very clearly marked in the disease progression, the initial emergence of tau or sudden acceleration of tau tangles could occur in multiple possible cell types within those brain regions.

We conclude that there is moderate overlap in genes associated with PD across GWAS, but very minimal overlap in genomic risk loci and cell types across GWAS. Our data therefore supports our initial hypothesis that PD GWAS are non-concordant in their identification of cell types associated with Parkinson's disease; additionally, our list of cell types associated with PD did include some oligodendrocyte and oligodendrocyte progenitor cell types, supporting previous associations of oligodendrocyte glia with PD.

Our limitations for this study include a lack of large PD GWAS, poor naming standards of scRNAseq cell types, and the complicated nature of Parkinson's and other tauopathies. However, the strength of this study is that the pipeline for comparison of multiple GWAS for the same trait could be extended to other neurological disorders similar to Parkinson's, such as insomnia, in order to further investigate other cell types that could contribute to PD. Furthermore, while the cell types associated with PD are mostly non-concordant across all PD GWAS, these PD-associated cell types were identified in previous literature and therefore our study helps to further validate the association of monoaminergic cells and oligodendrocyte glia with PD, while also providing a few new cell types, such as NF3 cortical cells or Vglut cells, to be studied in future investigations of Parkinson's disease.

Research in Context

Previously, cell types have been mapped to GWAS traits by identifying cell types that are enriched with the genes found in genomic risk loci identified in individual GWAS. Genomic risk loci contain multiple genes, however, and finding causal genes and subsequently causal cell types for a disease trait is a challenge. Therefore, we aim to provide a tool to map cell types from multiple GWAS rather than a single GWAS to identify overlapping cell types, with the assumption that cell types that repeatedly map to a disease trait are more consistently expressing the genes associated with that disease trait. We link glutamatergic (Vglut) cells, monoaminergic cells, and oligodendrocyte glial cell types with the Parkinson's disease trait.

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Supplementary Tables

GWAS	Nalls	Blauwendraat	Pankratz	Simon
Year	2019	2019	2012	2009
N_total	482730	34498	8477	5691
N_Case	33674	17996	4238	1713
N_Control	449056	16502	4239	3978
Percent_F_of_Case_and_Control	35.5 and NA	39.3 and NA	35.8 and 55.9	35.9 and 44.1
Age	21 - 84	20 - 89	44 - 77	20 - 87
Reference_Population	UK BioBank EUR	UK BioBank EUR	1000G Phase3	1000G Phase3
N_SNPs_Analyzed	17,510,618	6,813,423	98,200	453,217
N_GenomicRiskLoci	24	1	2	4
N_LeadSNPs	32	2	3	4
N_IndSigSNPs	83	2	3	5
N_CandidateSNPs	7694	119	3923	2920
N_GWASTagged_CandidateSNPs	4703	97	32	23

Supplementary Table 1: This table lists all the GWAS used in analysis for this study. Details about each study are also listed in the table.

	rsID	chr	p o s	p	sta rt	end	LeadSNPs	GWA S
1	rs14728 8664	1	152192 927	6.326e -14	151832 302	152277 663	rs147288664	Nalls
2	rs35749 011	1	155135 036	5.022e -30	154913 723	156154 860	rs35749011;rs35643925	Nalls
3	rs82310 6	1	205656 453	4.100e -10	205642 390	205757 824	rs823106;rs823116	Nalls
4	rs67410 07	2	135537 119	2.087e -12	135343 934	135583 585	rs6741007	Nalls
5	rs46132 39	2	169119 609	6.215e -13	169091 942	169159 689	rs4613239	Nalls
6	rs44888 03	3	582183 52	1.076e -08	582183 52	582259 74	rs4488803	Nalls
7	rs10513 789	3	182760 073	3.185e -13	182704 808	182860 042	rs10513789	Nalls
8	rs34311 866	4	951947	7.974e -23	843630	991192	rs34311866	Nalls
9	rs46984 12	4	157373 48	7.049e -14	155980 51	157445 76	rs4698412	Nalls

10	rs76957 20	4	771833 00	1.528e -09	771323 60	772194 02	rs7695720	Nalls
11	rs35620 3	4	906660 41	3.007e -41	905135 19	913079 91	rs79956144;rs356203;rs104339 53;rs1372518	Nalls
12	rs75646 569	5	603454 24	5.618e -13	600116 36	604939 44	rs4699965;rs75646569	Nalls
13	rs35265 698	6	325613 34	3.927e -11	323450 07	326133 60	rs35265698	Nalls
14	rs85829 5	7	232455 69	3.831e -09	231100 19	234072 40	rs858295	Nalls
15	rs62049 0	8	166975 79	6.456e -10	166954 18	167391 27	rs620490	Nalls
16	rs14481 4361	10	121410 917	9.065e -11	121295 884	121710 488	rs144814361	Nalls
17	rs32964 7	11	133764 666	1.944e -10	133764 405	133819 417	rs329647	Nalls
18	rs34637 584	12	407342 02	1.329e -12	403859 10		rs1491942;rs34637584;rs75505 347	Nalls
19	rs10847 864	12	123326 598	9.812e -13	123296 294	123326 598	rs10847864	Nalls
20	rs47744 17	15	619937 02	4.626e -08	619914 78	619975 99	rs4774417	Nalls

21	rs12934 900	16	309236 02	4.331e -11	308011 83	311554 58	rs12934900	Nalls
22	rs45662 08	17	160109 20	3.884e -08	158732 75	162255 06	rs4566208	Nalls
23	rs58879 558	17	440954 67	1.363e -21	434601 81	448634 13	rs58879558	Nalls
24	rs45880 66	18	406729 64	4.453e -09	406729 64	407415 35	rs4588066	Nalls
25	rs35620 3	4	906660	2.348e -09	906197	907881 96	rs356203;rs983361	Blauwer draat
26	rs11248 060	4	964359	1.948e -09	843674	991192	rs11248060	Pankrat
27	rs35621 9	4	906376 01	2.270e -21	906281 27	908089	rs356219;rs356198	Pankrat
28	rs31298 82	6	324095 30	1.208e -08	323428 22	324095 30	rs3129882	Pankrat
29	rs19952 8	17	448431 36	1.382e -11	434601 81	448634 13	rs3946526;rs199528	Pankrat
30	rs27369 90	4	906785 41	5.690e -09	906071	907797	rs2736990	Simon
31	rs37848 47	16	619774 49	1.660e -09	619481 50	619903 59	rs3784847	Simon

32	rs48897 30	17	217177 27	2.830e -11	217026 66	217396 44	rs4889730	Simon
33	rs41543 0	17	448591 44	4.501e -08	438596 40	448656 03	rs415430	Simon

Supplementary Table 2: This table lists all the genomic risk loci identified in the four GWAS used in this study. Chromosomal start and end positions are listed for each locus, as well as the lead SNP in the locus and the GWAS the locus was identified with.

Species	Dataset	Numbe r of Cells	Number of Genes	Brain Region of scRNAseq Dataset
Mouse	Linnarsson_GSE101601_M ouse_Somatosensory_corte x	2192	15826	somatosensory cortex
Mouse	Linnarsson_GSE59739_Mo use_Dorsal_root_ganglion	622	15084	dorsal root ganglion
Mouse	Linnarsson_GSE60361_Mo use_Cortex_Hippocampus	3005	15161	cortex and hippocampus
Mouse	Linnarsson_GSE75330_Mo use_Oligodendrocytes	5069	15816	oligodendrocytes
Mouse	Linnarsson_GSE78845_Mo use_Ganglia	213	13804	stellate/thoracic sympathetic glia
Mouse	Linnarsson_GSE104323_M ouse_Dentate_gyrus	24185	16146	dentate gyrus
Mouse	Linnarsson_GSE74672_Mo use_Hypothalamus	2881	15826	hypothalamus
Mouse	Linnarsson_GSE103840_M ouse_Dorsal_horn	1039	15116	dorsal horn
Human	Linnarsson_GSE101601_Hu man_Temporal_cortex	2028	21459	temporal cortex
Mouse	Linnarsson_GSE76381_Mo use_Midbrain	1695	16885	ventral midbrain
Mouse	GSE87544_Mouse_Hypotha	1039	15116	hypothalamus

	lamus			
Mouse	GSE82187_Mouse_Striatum	705	14189	striatum
Mouse	GSE93374_Mouse_Arc_ME	20921	1433	arcuate median complex
Mouse	GSE92332_Mouse_Epitheli um_droplet	7216	12865	small intestine
Mouse	GSE89164_Mouse_Hindbrai	4366	13176	mouse hindbrain
Human	GSE67835_Human_Cortex	466	19749	cortex
	GSE_98816_Brain_Vascular	3186	15302	vascular cells
Mouse	GSE84133_Mouse_Pancrea s	1886	12741	pancreas
Mouse	GSE106707_Mouse_Striatu m_Cortex	3417	8940	striatum and cortex
Human	GSE84133_Human_Pancre as	7266	19546	pancreas
Human	GSE104276_Human_Prefrontal_cortex_all_ages	2309	21177	PFC
Human?	PBMC_10x_68k	68579	32738	peripheral blood monovascular cells
Human	Allen_Human_MTG	15603	29115	MTG
Human	Allen_Human_LGN	1553	29115	LGN
Mouse	Allen_Mouse_ALM2	9573	16093	ALM2
Mouse	Allen_Mouse_LGd2	1874	16093	LGd2
Mouse	Allen_Mouse_VISp2	14249	16093	VISp2
Human	Dro_Nc_Human_Hippocam pus	14137	31852	hippocampus
Mouse	Dro_Nc_Mouse_Hippocamp us	11148	13335	hippocampus
Human?	PsychENCODE_Adult	27380	16243	total brain
	Total Number of Cells:	259802		

Supplementary Table 3: This table lists the 31 single-cell RNA sequencing (scRNAseq) datasets used in the mapping cell-type associations with Parkinson's disease GWAS with FUMA's Cell Type tool. This table lists the name of the scRNAseq dataset as listed on the FUMA platform. This table also provides information on how many cells were sequenced for each dataset, as well as how many genes were profiled in each dataset for the given cell types. There were a total of 259,802 cells contained in all 31 scRNAseq datasets.